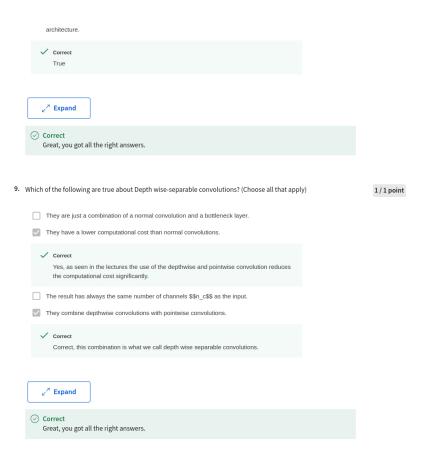
## Congratulations! You passed!

Grade received 80% Latest Submission Grade 80%

To pass 80% or higher Go to next item

1.	Which of the following do you typically see in ConvNet? (Check all that apply.)	0/1 point
	Multiple FC layers followed by a CONV layer.	
	Use of multiple POOL layers followed by a CONV layer.	
	ConvNet makes exclusive use of CONV layers.	
	Use of FC layers after flattening the volume to generate output classes.	
	∠ <sup>n</sup> Expand	
	Incorrect     No, this is not a common practice.	
2.	In order to be able to build very deep networks, we usually only use pooling layers to downsize the height/width of the activation volumes while convolutions are used with "valid" padding. Otherwise, we would downsize the input of the model too quickly.	1/1 point
	○ True	
	False	
	∠ <sup>7</sup> Expand	
	⊙ Correct Correct!	
3.	The motivation of Residual Networks is that very deep networks are so good at fitting complex functions that when training them we almost always overfit the training data. True/False?	1/1 point
	False	
	○ True	
	∠ <sup>™</sup> Expand	
	<ul> <li>Correct         Correct, very deep neural networks are hard to train and a deeper network does not always imply lower             training error. Residual Networks allow us to train very deep neural networks.     </li> </ul>	
4.	Which of the following equations captures the computations in a ResNet block?b	1/1 point
	$\bigcap_{a^{[l+2]}=g\left(W^{[l+2]}g\left(W^{[l+1]}a^{[l]}+b^{[l+1]}\right)+b^{[l+2]}\right)}$	
	$ \$\$a^{[[+2]]} = g Vefi( W^{[i+2]]} \setminus g Vefi( W^{[i+1]]} \cdot a^{[i]]} + b^{[i+1]} \cdot Vight) + b^{[i+2]} + a^{[i]} Vight) + a^{[i+1]} \times a^{[i+1]} $	
	$\frac{\$\$a\triangle(II+2)!-\alpha   laft MA(II+2)! \setminus a   laft MA(II+1)! \setminus a^{[[]} + b^{[[+1]} \cdot light) + b^{[[+2]} \cdot light) + lading [MathJax/jax/output/CommonHTML/jax,js]}{} \\ Loading [MathJax/jax/output/CommonHTML/jax,js]$	
	∠ <sup>n</sup> Expand	
	$\odot$ Correct Correct. This expresses the computations of a ResNet block, where the last term $a^{[l]}$ is the shortcut connection.	

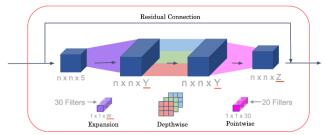
5. Which ones of the following statements on Residual Networks are true? (Check all that apply.)	1/1 point
✓ The skip-connection makes it easy for the network to learn an identity mapping between the input and the computation has Poster block.   The skip-connection makes it easy for the network to learn an identity mapping between the input and the computation in the Poster block.  The skip-connection makes it easy for the network to learn an identity mapping between the input and the computation.  The skip-connection makes it easy for the network to learn an identity mapping between the input and the computation.  The skip-connection makes it easy for the network to learn an identity mapping between the input and the computation.  The skip-connection makes it easy for the network to learn an identity mapping between the input and the computation is the poster block.  The skip-connection is the poster block bl	
input and the output within the ResNet block.  ✓ Correct	
This is true.	
$\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ $	
The skip-connections compute a complex non-linear function of the input to pass to a deeper layer in the network.	
Using a skip-connection helps the gradient to backpropagate and thus helps you to train deeper networks	
✓ Correct	
Typesetting math: 100%	
∠ <sup>7</sup> Expand	
⊘ Correct     Great, you got all the right answers.	
1  imes 1 convolutions are the same as multiplying by a single number. True/False?	0 / 1 point
○ False	
① True	
∠ <sup>7</sup> Expand	
⊗ Incorrect	
No, a $1  imes 1$ layer doesn't act as a single number because it makes a sum over the depth of the volume.	
6. Which of the following are true about the inception Network? (Check all that apply)	1/1 point
One problem with simply stacking up several layers is the computational cost of it.	
Correct Correct. That is why the bottleneck layer is used to reduce the computational cost.	
Making an inception network deeper won't hurt the training set performance.	
Inception blocks allow the use of a combination of 1x1, 3x3, 5x5 convolutions, and pooling by applying one layer after the other.	
Inception blocks allow the use of a combination of 1x1, 3x3, 5x5 convolutions and pooling by stacking up all the activations resulting from each type of layer.	
✓ Correct	
Correct. The use of several different types of layers and stacking up the results to get a single volume is at the heart of the inception network.	
∠ <sup>n</sup> Expand	
○ Correct	
Great, you got all the right answers.	
<ol> <li>Which of the following are common reasons for using open-source implementations of ConvNets (both the model and/or weights)? Check all that apply.</li> </ol>	1/1 point
Parameters trained for one computer vision task are often useful as pre-training for other	
computer vision tasks.	
✓ Correct True	
The same techniques for winning computer vision competitions, such as using multiple crops at test time, are widely used in practical deployments (or production system	
deployments) of ConvNets.  A model trained for one computer vision task can usually be used to perform data	
augmentation for a different computer vision task.	
✓ It is a convenient way to get working with an implementation of a complex ConvNet	



10. Fill in the missing dimensions shown in the image below (marked W, Y, Z).

1/1 point

## MobileNet v2 Bottleneck



W = 5, Y = 30, Z = 20
 W = 30, Y = 30, Z = 5
 W = 5, Y = 20, Z = 5
 W = 30, Y = 20, Z = 20

∠<sup>n</sup> Expand

**⊘** Correct